# Basic Perceptron

#### Trevor Davis Ducharme

October 9, 2024

### 1 Introduction

This document provides a mathematical overview of the basic perceptron model.

# 2 Perceptron Model

A perceptron is a simple linear binary classifier. The output of the perceptron is given by:

$$z = \left(\sum_{i=1}^{n} w_i \cdot x_i\right) + b$$
$$y = A(z)$$

where:

- $\bullet$  A is the activation function
- $\bullet$  w is the weight vector
- $\bullet$  **x** is the input vector
- $\bullet$  b is the bias term

# 3 Training the Perceptron

The perceptron is trained using the following equations:

#### 3.1 Necessary Derivatives

$$\frac{\partial y}{\partial z} = A'(z)$$
$$\frac{\partial z}{\partial \mathbf{w_i}} = \mathbf{x_i}$$
$$\frac{\partial z}{\partial b} = 1$$

### 3.2 Updating the Weights

The weights are updated as follows:

$$\mathbf{w} \leftarrow \mathbf{w} + \Delta \mathbf{w}$$

where:

$$\Delta \mathbf{w_i} = -\eta \cdot L'(y, \hat{y}) \cdot \frac{\partial y}{\partial z} \cdot \frac{\partial z}{\partial \mathbf{w_i}}$$

Here,  $\eta$  is the learning rate, L is the loss function, and  $\hat{y}$  is the target output.

## 3.3 Updating the Bias

The bias term is updated as follows:

$$\mathbf{b} \leftarrow \mathbf{b} + \Delta \mathbf{b}$$

where:

$$\Delta \mathbf{b} = - \boldsymbol{\eta} \cdot L'(\boldsymbol{y}, \hat{\boldsymbol{y}}) \cdot \frac{\partial \boldsymbol{y}}{\partial \boldsymbol{z}} \cdot \frac{\partial \boldsymbol{z}}{\partial \boldsymbol{b}}$$